# Statistics: Lecture 1 - Probability Theory and Random Variables

#### Jan Bauer

jan.bauer@dhbw-mannheim.de

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#### Vocabs

- simple events/outcomes  $\omega_1, \omega_2, \ldots$ : Relevant outcomes for a random phenomena
- sample space  $\Omega$ : Set of all simple outcomes
- (random) experiment: Procedure having a defined set of possible outcomes
- event: Subset of the sample space (all events: All subsets)
- An event is said to occur when one of the outcomes it contains occurs
- complementary event to the event  $E_i$ : Set of outcomes not contained in  $E_i$  (denoted by  $E_i^c$ )

#### Example: Coin Toss

• experiment: Tossing a coin

• simple outcomes:  $\omega_1 = H$  (head) and  $\omega_2 = T$  (tail)

• sample space:

$$\Omega = \{H,T\}$$

• events:  $E_1 = H$  and  $E_2 = T$ 



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# Example: Tossing a Coin twice

• experiment: Tossing a coin twice

- basic outcomes:  $\omega_1 = HH$ ,  $\omega_2 = HT$ ,  $\omega_3 = TH$  and  $\omega_4 = TT$ ,
- sample space:

$$\Omega = \{HH, HT, TH, TT\}$$

• (possible) events:

$$E_1 = \{HH, HT, TH\}$$
 ("at least one head")  
 $E_2 = \{HH, TT\}$  ("both faces same")

$$E_1^c = \{TT\}$$
 ("no head") (complement of  $E_1$ )





# Vocabs (classic algebra of sets)

- certain event:  $\Omega$
- impossible event: Ø
- union of A and B: Set of sample points that belong to at least A or B (denoted by  $A \cup B$ )
- intersection of A and B: Set of sample points that belong to both A and B (denoted by  $A \cap B$ )

Example (previous)

• 
$$E_1 \cup E_2 = \Omega$$

$$\bullet E_1 \cap E_2 = \{HH\}$$



### Probability

• Probability is a value assigned to a simple outcome.

Consider  $\Omega = {\omega_1, \ldots, \omega_n}$ . To each  $\omega_i$ , we assign its probability  $P(\omega_i) = p_i$ , which satisfies

$$p_i \in [0,1] \text{ and } \sum_{i=1}^n p_i = 1$$
.

For an event A, we assign its probability

$$P(A) = \sum_{\omega_i \in A} p_i .$$

 $\rightarrow$  the probability of an event A is the sum of the probabilities of the simple outcomes contained in this event.



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# Probability (Example)

Tossing a coin twice:

• 
$$P(\omega_1) = P(HH) = p_1 = \frac{1}{4}$$
,  
 $P(\omega_2) = P(HT) = p_2 = \frac{1}{4}$ ,  
 $P(\omega_3) = P(TH) = p_3 = \frac{1}{4}$ ,  
 $P(\omega_4) = P(TT) = p_4 = \frac{1}{4}$  with

$$\sum_{i=1}^{4} p_i = \frac{1}{4} + \frac{1}{4} + \frac{1}{4} + \frac{1}{4} = 1 .$$

•  $E_1 = \{HH, HT, TH\}$  ("at least one head")

$$\rightarrow$$
 P(E<sub>1</sub>) = p(HH) + p(HT) + p(TH) =  $\frac{1}{4} + \frac{1}{4} + \frac{1}{4} = \frac{3}{4}$ 

•  $E_1^c = \{TT\}$  ("no head")

$$\rightarrow$$
  $P(E_1^c) = 1 - P(E_1) = 1 - \frac{3}{4} = \frac{1}{4}$ 

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(probability of a complementary event:  $P(E) = 1 - P(E^c)$ 

#### Kolmogorov Axioms

Let  $\mathcal{E}$  be the event space containing all events. For the events  $E_1, E_2, \ldots$  in  $\mathcal{E}$  and the sample space  $\Omega$  it holds that

non-negativity:

$$P(E_i) \ge 0 \quad \forall E_i \in \mathcal{E}$$

• unit measure:

$$P(\Omega) = 1$$

•  $\sigma$ -additivity: If  $E_1, E_2, \ldots$  are disjoint  $(E_i \cap E_j = \emptyset, j \neq i)$ , then

$$P\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} P(E_i)$$





# Consequences of Kolmogorov's Axioms

For the events  $E_1, E_2, \ldots$  in the event space  $\mathcal{E}$  and the sample space  $\Omega$  it holds that

•  $P(\emptyset) = 0$ 

- if  $E_i \cap E_j = \emptyset \Rightarrow P(E_i \cap E_j) = 0$
- $P(E_i) \in [0,1] \quad \forall E_i \in \mathcal{E}$
- $P(E_i) = 1 P(E_i^c) \quad \forall E_i \in \mathcal{E}$
- if  $E_i \subseteq E_j \Rightarrow P(E_i) \le P(E_j)$





#### The Addition Rule

• For any two events A and B, it holds that

$$P(A \cup B) = P(A) + P(B) - P(A \cap B) .$$

• For the events  $E_1, E_2, \ldots$ , it holds that

$$P\left(\bigcup_{i} E_{i}\right) \leq \sum_{i} P(E_{i})$$
.

(note the difference to  $\sigma$ -additivity)



#### The Addition Rule (Example)

Example: Roll a die once. What is the probability, that the number is even or that the number is smaller than four?

$$\leadsto A = \{2, 4, 6\} \text{ and } B = \{1, 2, 3\}.$$
 Then

$$P(A \cup B)$$
=  $P(A) + P(B) - P(A \cap B)$   
=  $P(\{2, 4, 6\}) + P(\{1, 2, 3\}) - P(\{2\})$   
=  $\frac{3}{6} + \frac{3}{6} - \frac{1}{6}$   
=  $\frac{5}{6}$ 



#### Probability Assignment: Probability Trees

• Used to represent a series of events.

ullet Example: Tossing a coin twice. A is the first toss, B the second one.



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# Probability Assignment: Probability Tables (two Events)

Example: Tossing a coin twice. A is the first toss, B is the second toss.

Then the probability table providing the joint probabilities is given by

	A = H	A = T	P(B)
B = H	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$
B = T	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$
P(A)	$\frac{1}{2}$	$\frac{1}{2}$	1

• 
$$P(A = H \cap B = H) = \frac{1}{4}$$

• 
$$P({A = H \cap B = H}) \cup {A = T \cap B = T})$$
  
=  $P(A = H \cap B = H) + P(A = T \cap B = T)$   
=  $\frac{1}{4} + \frac{1}{4} = \frac{1}{2}$ 

since  $\{A = H \cap B = H\} \cap \{A = T \cap B = T\} = \emptyset$  are disjoint sets



# Probability Assignment: Probability Tables (three Events)

Example: Tossing a coin three times. A is the first toss, B is the second toss and C is the third toss, respectively. Then the probability table providing the joint probabilities is given by

C = H	A = H	A = T	$P(B \cap C = H)$
B = H	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{4}$
B = T	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{4}$
$P(A \cap C = H)$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$

C = T	A = H	A = T	$P(B \cap C = T)$
B = H	1/8	$\frac{1}{8}$	$\frac{1}{4}$
B = T	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{4}$
$D(A \circ C \cap T)$	1	1	40.10.45



#### Conditional Probability

• What is the probability of me going to a party tonight?

• What is the probability of me going to party tonight, when there is beer for free?



#### Conditional Probability

 Given two events A and B, the conditional probability of A given B is defined as

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$
,  $P(B) > 0$ .

• Equivalently, it holds that

$$P(A \cap B) = P(A|B) \cdot P(B)$$
.





### Conditional Probability (Example)

Consider the event P with outcomes y and n, which corresponds to the outcome of going to the party (yay) or not (nay). Further, we have the event of beer for free, B, which has the outcomes y and n. We assume the following probability table:

	P = y	P = n	P(B)
B = y	$\frac{1}{2}$	$\frac{1}{8}$	<u>5</u> 8
B = n	0	$\frac{3}{8}$	$\frac{3}{8}$
P(P)	$\frac{1}{2}$	$\frac{1}{2}$	1

What is the probability of me going to party tonight, when there is beer for free?

$$\mathrm{P}(P=y|B=y) = \frac{\mathrm{P}(P=y\cap B=y)}{\mathrm{P}(B=y)} = \frac{1/2}{5/8} = \frac{4}{5} \; .$$



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#### The Total Probability Rule

Let  $E_1, E_2, \ldots$  be disjoint events with  $\bigcup_i E_i = \Omega$  and suppose that  $P(E_i) > 0 \ \forall i$ . Let A be an arbitrary event, then

$$P(A) = \sum_{i} P(E_i)P(A|E_i)$$
.



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#### Bayes' Theorem

Let  $E_1, E_2, ...$  be events and suppose that  $P(E_i) > 0 \,\forall i$ . Let A be an arbitrary event satisfying P(A) > 0, then

$$P(E_i|A) = \frac{P(E_i)P(A|E_i)}{P(A)} \stackrel{(TPR)}{=} \frac{P(E_i)P(A|E_i)}{\sum_i P(E_i)P(A|E_i)},$$

where the last conclusion follows if the Total Probability Rule assumptions are satisfied.



#### Independence

• What is the probability of me going to a party tonight?

• What is the probability of me going to party to night, when there is bad weather in England?



#### Independence

• Two events A and B are said to be independent, if

$$P(A|B) = P(A) \quad (\text{or } P(B|A) = P(B)) .$$

• Equivalently, it holds that

$$P(A \cap B) = P(A) \cdot P(B)$$
.

• Where does this come from?

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = P(A) \Leftrightarrow P(A \cap B) = P(A) \cdot P(B)$$
.





#### Independence of more than two events

• The events  $E_1, E_2, \ldots$  are said to be independent, if

$$P\left(\bigcap_{i} E_{i}\right) = \prod_{i} P(E_{i})$$
.



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#### Check For Independence (Example)

Consider again the following probability table:

	P = y	P = n	P(B)
B = y	$\frac{1}{2}$	<u>1</u> 8	<u>5</u> 8
B = n	0	$\frac{3}{8}$	$\frac{3}{8}$
P(P)	$\frac{1}{2}$	$\frac{1}{2}$	1

Are P and B independent? Check  $P(P \cap B) = P(P) \cdot P(B)$  for all outcomes y and n:

$$P(P = y \cap B = y) = \frac{1}{2} \neq \frac{5}{16} = P(P = y) \cdot P(B = y)$$

 $\leadsto$  not independent



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#### Random Variables (Introduction)

 We want to consider random experiments associated with numerical outcomes

• Example: Rolling a die  $\rightsquigarrow$  outcomes 1, 2, 3, 4, 5 and 6

- Example: Tossing a coin → translate outcomes "H" and "T" to 1 and 0
- Random variables are the base of probability theory We will only focus on the application, not the theory
- The definitions are not always mathematically complete; although sufficient for this course

# Random Variable (definition)

• Our (sufficient) definition of a random variable (rv):

A random variable X is a function

$$X: \Omega \to \mathbb{R}, \ \omega \mapsto x = X(\omega)$$

from a sample space,  $\Omega$ , to the real numbers,  $\mathbb{R}$ .

- X is called a discrete rv if it is defined over a sample space having finite or a countably infinite number of sample points. In this case, X takes on discrete values.
- X is called a continuous rv if it is distributed over one or more continuous intervals in R.



### Random Variable (Example: Coin Toss)

• sample space:

$$\Omega = \{H, T\}$$

• we want to build a RV X that assigns the value 1 for "Head" and 0 for "Tail":

$$X(\omega) = \begin{cases} 1 & \text{if } \omega = H \\ 0 & \text{if } \omega = T \end{cases}$$



#### Random Variable (Example: Rolling a die)

sample space:

$$\Omega = \{"1", "2", "3", "4", "5", "6"\}$$

we want to build a RV X that assigns the value 1 when the die shows "1", the value 2 when the die shows "2",...

$$X(\omega) = \begin{cases} 1 & \text{if } \omega = \text{"1"} \\ 2 & \text{if } \omega = \text{"2"} \\ 3 & \text{if } \omega = \text{"3"} \\ \vdots & \vdots \end{cases}$$



### Random Variable (Example: Rolling a die "reverse")

• sample space:

$$\Omega = \{"1", "2", "3", "4", "5", "6"\}$$

• we want to build a RV X that assigns the value 6 when the die shows "1", the value 5 when the die shows "2",...

$$X(\omega) = \begin{cases} 6 & \text{if } \omega = \text{"1"} \\ 5 & \text{if } \omega = \text{"2"} \\ 4 & \text{if } \omega = \text{"3"} \\ \vdots & \vdots \end{cases}$$



# Random Variable (Properties)

Let X and Y be two random variables. Then

- X + Y is a RV
- X Y is a RV
- $c \cdot X$  is a RV (for  $c \in \mathbb{R}$  a constant)
- $X^2$  is a RV
- $X \cdot Y$  is a RV
- $\frac{1}{X}$  is a RV (if  $X \neq 0$ )
- $\frac{X}{Y}$  is a RV (if  $Y \neq 0$ )

